

**ROBUST PERCENTILE BOOTSTRAP TEST WITH
MODIFIED ONE-STEP M -ESTIMATOR (MOM): AN
ALTERNATIVE MODERN STATISTICAL ANALYSIS**

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Abstrak

Kenormalan dan homoskedastisiti merupakan dua andaian utama yang perlu dipenuhi apabila berurusan dengan ujian-ujian parameter klasik untuk perbandingan kumpulan. Pelanggaran mana-mana andaian tersebut akan menyebabkan keputusan ujian menjadi tidak sah. Walau bagaimanapun, pada realitinya, kedua-dua andaian tersebut sukar dicapai. Untuk mengatasi masalah tersebut, kajian ini mencadangkan pengubahsuaian satu kaedah yang dikenali sebagai ujian Bootstrap Berparameter dengan menggantikan min sebenar, \bar{X} dengan ukuran lokasi yang sangat teguh iaitu penganggar-M satu-langkah terubahsuai (*MOM*). (*MOM*) merupakan min terpangkas tidak simetri. Penggantian ini akan menjadikan ujian Bootstrap Berparameter lebih teguh untuk perbandingan kumpulan. Dalam kajian ini, kriteria pemangkasan untuk *MOM* menggunakan dua penganggar skala yang amat teguh iaitu MAD_n dan T_n . Satu kajian simulasi telah dijalankan untuk mengkaji prestasi kaedah yang dicadangkan berdasarkan kadar Ralat Jenis I. Untuk mengenal pasti kekuatan dan kelemahan kaedah, lima pembolehubah iaitu: bilangan kumpulan, saiz sampel seimbang dan tak seimbang, jenis taburan, keheterogenan varians, dan sifat pasangan bagi saiz sampel dan varians kumpulan dimanipulasi untuk menghasilkan pelbagai keadaan yang biasanya wujud dalam kehidupan sebenar. Prestasi kaedah yang dicadangkan kemudiannya dibandingkan dengan ujian parameter klasik dan ujian tidak berparameter yang paling kerap digunakan untuk dua (ujian-*t* tidak bersandar dan ujian *Mann Whitney* masing-masing) dan lebih daripada dua kumpulan tidak bersandar (ANOVA dan ujian *Kruskal Wallis* masing-masing). Dapatan kajian menunjukkan bahawa, untuk dua kumpulan, ujian Bootstrap Berparameter yang teguh menunjukkan prestasi yang baik di bawah keadaan varians heterogen dengan taburan normal atau taburan terpencong. Manakala untuk lebih daripada dua kumpulan, ujian tersebut menjana pengawalan Ralat Jenis I yang baik di bawah varians heterogen dan taburan terpencong. Dalam perbandingan dengan kaedah parameter klasik dan kaedah tidak berparameter, ujian yang dicadangkan menunjukkan prestasi yang lebih baik di bawah taburan terpencong dan varians heterogen. Prestasi setiap prosedur juga ditunjukkan dengan menggunakan data sebenar. Secara umumnya, prestasi Ralat Jenis I bagi ujian yang dicadangkan adalah sangat menyakinkan walaupun andaian kenormalan dan homoskedastisiti dilanggar.

Kata kunci: Titik kegagalan, Heterogen, Taburan terpencong, Ralat Jenis I.

Abstract

Normality and homoscedasticity are two main assumptions that must be fulfilled when dealing with classical parametric tests for comparing groups. Any violation of the assumptions will cause the results to be invalid. However, in reality, these assumptions are hardly achieved. To overcome such problem, this study proposed to modify a method known as Parametric Bootstrap test by substituting the usual mean, \bar{X} with a highly robust location measure, modified one step M-estimator (*MOM*). *MOM* is an asymmetric trimmed mean. The substitution will make the Parametric Bootstrap test more robust for comparing groups. For this study, the trimming criteria for *MOM* employed two highly robust scale estimators namely MAD_n and T_n . A simulation study was conducted to investigate on the performance of the proposed method based on Type I error rates. To highlight the strength and weakness of the method, five variables: number of groups, balanced and unbalanced sample sizes, types of distributions, variances heterogeneity and nature of pairings of sample sizes and group variances were manipulated to create various conditions which are common to real life situations. The performance of the proposed method was then compared with the most frequently used parametric and non parametric tests for two (independent sample *t*-test and Mann Whitney respectively) and more than two independent groups (ANOVA and Kruskal Wallis respectively). The finding of this study indicated that, for two groups, the robust Parametric Bootstrap test performed reasonably well under the conditions of heterogeneous variances with normal or skewed distributions. While for more than two groups, the test generate good Type I error control under heterogeneous variances and skewed distributions. In comparison with the parametric and non parametric methods, the proposed test outperforms its counterparts under non-normal distribution and heterogeneous variances. The performance of each procedure was also demonstrated using real data. In general, the performance of Type I error for the proposed test is very convincing even when the assumptions of normality and homoscedasticity are violated.

Keywords: Breakdown point, Heterogeneity, Skewed distributions, Type I error.

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List of Abbreviations

ANOVA	Analysis of variance
Parametric Bootstrap	A statistical method for testing the equality of central tendency
MAD_n	Median absolute deviation about median
T_n	A scale estimator

List of Publications

- Md Yusof, Z., Harun, N. H., Syed Yahaya, S. S. & Abdullah, S. (2013). A modified parametric bootstrap: an alternative to classical parametric test. *In proceeding of the World Conference on Integration of Knowledge 2013*, 25 – 26 November, Langkawi, Malaysia.
- Harun, N. H, Md Yusof, Z. (2013). Testing the Equality of Central Tendency using Robust Parametric Bootstrap Test with *MOM* Estimator for Two Groups Case. *In proceeding of the 1st Innovation and Analytics Conference and Exhibition 2013*, 29 December, Universiti Utara Malaysia, Malaysia.
- Harun, N. H, Md Yusof, Z. (2014). Robust Parametric Bootstrap Test with *MOM* Estimator: An Alternative to Independent Sample *t*-Test. In proceeding of the *3rd International Conference on Quantitative Sciences and Its Applications 2014*, 12 – 14 August, Langkawi, Malaysia.

CHAPTER ONE

BACKGROUND

1.1 Introduction

Statistics encompasses a wide variety of activities, ideas and results that can handle the situations involving uncertainties. Statistics consists of two basic statistical analysis namely descriptive statistics and inferential statistics. Recording and summarizing a data set is the main purpose of descriptive statistics whereas inferential statistics involves drawing conclusion and making decisions. There are extensive studies in testing equality of central tendency measures in inferential statistics using statistical method in order to make inferences based on obtained results. Basically, classical parametric tests such as analysis of variance (ANOVA) and independent sample t -test are often used in testing the central tendency measure by researchers rather than other methods since the aforementioned methods provide a good control of Type I error and generally more powerful than other methods when all the assumptions are fulfilled (Wilcox & Keselman, 2010).

ANOVA is used to determine the mean equality for more than two groups while independent samples t -test is used to determine the mean equality for two independent groups. However, a characteristic of these procedures is the fact that making inference depends on certain assumptions that need to be fulfilled. There are three main assumptions that need to be fulfilled before making inference on the classical parametric test such as: (a) collecting data from independent groups, (b) normally distributed data and (c) variances in the groups are equal (homoscedasticity). However, the specific interest of this study is to focus only for

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